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NORM PROPERTIES OF C-NUMERICAL RADII

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ABSTRACT. Given $n \times n$ complex matrices A, B, the C-numerical radius of A is the nonnegative quantity

 $r_{C}(A) \equiv \max\{|tr(CU^*AU)| : U \text{ unitary}\}.$

For C = diag(1,0,...,0) it reduces to the classical numerical radius $r(A) = max\{|x^*Ax| : x^*x = 1\}$. We show that r_C is a generalized matrix norm if and only if C is nonscalar and tr $C \neq 0$. Next, we consider an arbitrary generalized matrix norm and characterize all constants v > 0for which νN is multiplicative. A technique to obtain such ν is then applied to C-numerical radii with Hermitian C. In particular we find that vr is a matrix norm if and only if v > 4.

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1. Introduction

Let $C_{n\times n}$ be the algebra of $n\times n$ complex matrices and let V_n be its unitary group. Given A, C $\in C_{n\times n}$, the C-numerical range of A is the compact set

$$W_{C}(A) = \{tr(CU^*AU) : U \in \mathcal{U}_n\}.$$

This definition together with some properties of $W_{\mathbb{C}}(A)$ were presented by the authors in [2].

It is not hard to see (compare [2], Lemma 9), that $W_{\mathbb{C}}(A)$ is invariant under unitary similarities of A or C. Hence, if C is normal with eigenvalues γ_1 , we easily find that

$$(1.1) \quad \mathbf{W}_{\mathbf{C}}(\mathbf{A}) = \mathbf{W}_{\mathbf{diag}(\gamma_1, \dots, \gamma_n)}(\mathbf{A}) = \left\{ \sum_{j=1}^{n} \gamma_j \mathbf{x}_j^* \mathbf{A} \mathbf{x}_j : \{\mathbf{x}_j\} \in \Lambda_n \right\},$$

 Λ_n being the set of orthonormal bases for C_n . In particular, for C = diag(1,0,...,0), we obtain the classical range

$$W(A) = \{x^*Ax : x^*x = 1\}.$$

Associated with the classical range is the numerical radius

$$r(A) = \max\{|z| : z \in W(A)\}.$$

Similarly, we define the C-numerical radius to be

$$r_{C}(A) = \max\{|z| : z \in W_{C}(A)\}.$$

The main purpose of this work is to study the norm properties of r_{C} . The situation is trivial for n = 1, so without further reference we assume troughout the paper that n > 2.

White Section

We use the following standard definitions.

(i) A mapping $A \to N(A)$ is a <u>semi-norm</u> on $C_{n\times n}$, if for any and $\alpha \in C$, Dist. AVAIL and/or SPECIAL

 $N(A) \ge 0;$

 $N(\alpha A) = |\alpha|N(A);$

 $N(A + B) \leq N(A) + N(B)$.

(ii) A semi-norm is a generalized matrix norm if it is positive definite, that is,

N(A) > 0 for $A \neq 0$.

(iii) A generalized matrix norm is a matrix norm if it is (sub-) multiplicative, i.e., for all A, B,

 $N(AB) \leq N(A)N(B)$.

Without difficulty we obtain

THEOREM 1. For any C, r is a semi-norm.

The questions of definiteness and multiplicativity are much more complicated.

In Section 2 we characterize those C for which r_{C} is positive definite. We show that r_{C} is a generalized matrix norm if and only if C is not scalar and tr C \neq 0. This result agrees with the well known fact that the classical radius r is a generalized matrix norm.

The classical radius is not multiplicative, [4]. Hence, in general, a C-radius cannot be expected to be a matrix norm.

In Section 3 we consider arbitrary generalized matrix norms, and characterize all positive constants ν for which νN is multiplicative. A technique of finding such <u>multiplicativity factors</u> is given by a theorem of Gastinel [1].

The above technique (aided by some combinatorial inequalities

obtained in Section 4) is applied in Section 5 to find multiplicativity factors for C-numerical radii with Hermitian C. In particular we find that vr is a matrix norm if and only if $v \ge 4$.

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2. Norm Characterization of C-radii.

THEOREM 2. r_C is a generalized matrix norm if and only if

(2.1) C is nonscalar and tr
$$C \neq 0$$
.

In the proof we use the following three lemmas in which A, C are given $n \times n$ matrices.

LEMMA 1. Let m be an integer with 1 < m < n. If C leaves invariant all m-dimensional subspaces of \mathbb{C}^n , then C is scalar.

<u>Proof.</u> Since m < n, then each one-dimensional subspace of \mathbb{C}^n is an intersection of subspaces of dimension m, which by hypothesis, are fixed by C. This implies that C fixes all one-dimensional subspaces of \mathbb{C}_n .

Now let $\{e_j\}_{j=1}^n$ be the standard basis of \mathbb{C}^n . By the preceding argument, there exist scalars $\lambda_1,\ldots,\lambda_n$, μ , such that

$$Ce_j = \lambda_j e_j, \quad 1 \leq j \leq n,$$

and

$$C \sum_{j=1}^{n} e_{j} = \mu \sum_{j=1}^{n} e_{j}.$$

Hence, $\mu \Sigma e_j = \Sigma \lambda_j e_j$, and we conclude that $\lambda_j = \mu$, $1 \le j \le n$. Therefore,

$$Ce_j = \mu e_j, \quad 1 \leq j \leq n;$$

i.e., $C = \mu I$, and the lemma follows.

LEMMA 2. If

then either A or C are scalar.

<u>Proof.</u> Suppose A is not scalar and let us prove that C is. Let λ be an eigenvalue of A with corresponding eigenspace χ_{λ} of dimension m. Since A is not scalar, then

$$1 \le m = dim(\chi_{\lambda}) < dim(\zeta^n) = n.$$

Now, for arbitrary $U \in \mathcal{U}_h$, U^*AU also has λ as eigenvalue with corresponding eigenspace $U^*\mathcal{V}_{\lambda}$. Thus, for every vector $\mathbf{v} \in U^*\mathcal{V}_{\lambda}$,

$$U^*AU(Cv) = C(U^*AUv) = C(\lambda v) = \lambda(Cv).$$

It follows that

$$Cv \in U^*Y_{\lambda} \quad \forall v \in U^*Y_{\lambda}$$

that is, C leaves $U^*\chi_{\lambda}$ invariant. Since $\dim(\chi_{\lambda}) = m$ and U^* is arbitrary, we find that C leaves invariant all m-dimensional subspaces of ζ^n . Hence, by Lemma 1, C is scalar and the proof is complete.

LEMMA 3. If

then

<u>Proof.</u> Let S be skew-Hermitian; then $e^{\Theta S}$ is unitary for all real Θ , and so is $Ue^{\Theta S}$. By hypothesis therefore,

$$f(\theta) = tr[C(Ue^{\Theta S})*A(Ue^{\Theta S})] = constant, \quad \theta \in \mathbb{R};$$

and consequently,

$$\frac{d}{d\theta}f(\theta) = \frac{d}{d\theta}tr(Ce^{-\theta S}U^*AUe^{\theta S}) =$$

$$= tr(Ce^{-\theta S}U^*ASe^{\theta S} - CSe^{-\theta S}U^*AUe^{\theta S}) = 0.$$

Evaluating the derivative at $\theta = 0$ we obtain

hence for all skew-Hermitian S (and all unitary U),

$$tr[(CU*AU - U*AUC)S] = 0.$$

Since every matrix B is a linear combination of skew-Hermitians*, the last identity implies

$$tr[(CU^*AU - U^*AUC)B] = 0 \quad \forall B \in C_{n \times n}$$

Thus,

and the lemma is proven.

<u>Proof of Theorem 2.</u> By Theorem 1, it suffices to show that (2.1) holds if and only if r_c is positive definite.

If C is scalar, namely $C = \lambda I$, then any $A \neq 0$ with tr A = 0 gives

$$\mathbf{r}_{\mathbf{C}}(\mathbf{A}) = |\lambda \, \mathbf{tr} \, \mathbf{A}| = 0.$$

^{*}For example, $B = S_1 - 1S_2$ with $S_1 = \frac{1}{2}(B - B^*)$, $S_2 = \frac{1}{2}(B + B^*)$.

Also, if tr C = 0, then

$$r_{C}(I) = |tr C| = 0.$$

Thus, violation of (2.1) implies the indefiniteness of $r_{C}(\cdot)$.

Conversely, let (2.1) hold. If $r_{C}(A) = 0$, then by definition

$$tr(CU^*AU^*) = 0 \quad \forall U \in \mathcal{U}_n;$$

so by Lemma 3,

By Lemma 2, therefore, either C or A are scalar, and since C is not, A is. Setting $A = \mu I$ we have

$$r_{C}(A) = |\mu \text{ tr } C| = 0,$$

and since tr C \neq 0, then μ must vanish and the proof is established.

EXAMPLE 1. The k-numerical range, $1 \le k \le n$, was defined by Halmos [3, § 167] to be

 $W_k(A) = \{tr(PA) : P \text{ orthonormal projection of rank } k\}.$

We easily verify that

$$W_k(A) = W_{C_k}(A)$$
 where $C_k = I_k \oplus O_{n-k}$.

Thus, the k-numerical radius

$$r_k(A) = \max\{|z| : z \in W_k(A)\},$$

is a generalized matrix norm if and only if $1 \le k \le n - 1$. In particular $r(A) = r_1(A)$ is a generalized norm while $r_n(A) = |\operatorname{tr} A|$ is not.

Multiplicativity Factors and Gastinel's Theorem

Given a semi-norm N on $\mathbb{Q}_{n \times n}$ and a constant $\nu > 0$, then obviously

$$N_{\nu} \equiv \nu N$$

is a semi-norm too. Similarly, N is definite if and only if N_{ν} is. In any case the new norm may or may not be multiplicative. If it is, we say that ν is a <u>multiplicativity factor</u> of N.

A characterization of multiplicativity factors for generalized matrix norms is given in Theorem 4. We first prove, however, that indefinite nontrivial semi-norms have no multiplicativity factors.

THEOREM 3. An indefinite semi-norm N on $C_{n\times n}$ is multiplicative if and only if N \equiv 0.

<u>Proof.</u> The trivial semi-norm is certainly multiplicative. So let N be indefinite and multiplicative, and let us show that $N \equiv 0$.

Since N is indefinite, then N(A) = 0 for some $A \neq 0$. Let $\alpha_{\ell k}$ be a nonvanishing entry of A, and denote by E_{ij} the matrix whose (i,j) element is 1 and the others are zero. Since

then by multiplicativity,

$$|\alpha_{\ell_k}|_{N(E_{i,j})} = N(\alpha_{\ell_k}E_{i,j}) \le N(E_{i,\ell})N(A)N(E_{k,j}) = 0.$$

We conclude that

$$N(E_{i,j}) = 0 \quad \forall 1 \leq i, j \leq n;$$

thus for any $B = (\beta_{ij}) \in C_{n \times n}$,

$$N(B) = N\left(\sum_{i,j} \beta_{i,j} E_{i,j}\right) \leq \sum_{i,j} |\beta_{i,j}| N(E_{i,j}) = 0,$$

and the theorem follows.

THEOREM 4. If N is a generalized matrix norm, then v is a multiplicativity factor of N (i.e., N, is a matrix norm) if and only if

$$v \ge v_{N} \equiv \max_{A,B \neq 0} \frac{N(AB)}{N(A)N(B)}$$
.

Proof. We write v_{N} in the form

$$v_{N} = \max\{N(AB) : N(A) = N(B) = 1\},$$

and use a compactness argument to conclude that ν_N is well defined. It is clear then that $\nu_N > 0$.

Now, if $v \geq v_N$, then

$$N_{\nu}(AB) = \nu N(AB) \le \nu \nu_{N} N(A) N(B) \le \nu^{2} N(A) N(B) = N_{\nu}(A) N_{\nu}(B);$$

hence N is multiplicative.

Conversely, if ν satisfies $0 < \nu < \nu_N$, we can find matrices A, B such that $\nu N(A)N(B) < N(AB)$. Thus we have

$$N_{\nu}(AB) = \nu N(AB) > \nu^2 N(A)N(B) = N_{\nu}(A)N_{\nu}(B),$$

and the proof is complete.

As an immediate consequence we have established

COROLIARY 1. A generalized matrix norm N_V is a matrix norm if and only if $\nu_N \leq 1$.

In practice, Theorem 4 offers limited help since in general, v_N is

not easily evaluated. In the case of C-numerical radii, we were unable to find the optimal factor except for the classical radius.

An alternative way of finding multiplicativity factors is suggested by the following, somewhat stronger version of a theorem by Gastinel, [1].

THEOREM 5. Let N be a semi-norm, M a matrix norm, and $\eta \ge \xi > 0$ constants such that

$$(3.1) \qquad \qquad \underline{\xi}M(A) \leq N(A) \leq \eta M(A) \qquad \forall A \in \mathcal{Q}_{n \times n}.$$

Then,

- (i) N is a generalized matrix norm.
- (ii) For any $v \ge \eta/\xi^2$, N_v is a matrix norm.
- (iii) If $\eta/\xi^2 \le 1$, then N is a matrix norm.

<u>Proof.</u> Part (i) is trivial, and for part (ii) we should merely note that

$$\begin{split} \mathbf{N}_{\nu}(\mathbf{A}\mathbf{B}) &= \nu \mathbf{N}(\mathbf{A}\mathbf{B}) \leq \nu \eta \mathbf{M}(\mathbf{A}\mathbf{B}) \leq \nu \eta \mathbf{M}(\mathbf{A}) \mathbf{M}(\mathbf{B}) \\ &\leq \frac{\nu \eta}{\epsilon^2} \mathbf{N}(\mathbf{A}) \mathbf{N}(\mathbf{B}) \leq \nu^2 \mathbf{N}(\mathbf{A}) \mathbf{N}(\mathbf{B}) = \mathbf{N}_{\nu}(\mathbf{A}) \mathbf{N}_{\nu}(\mathbf{B}). \end{split}$$

Part (iii) then follows.

We recall, of course, that any two norms on $C_{n\times n}$ are equivalent. Thus if N of Theorem 5 is known to be a matrix norm, then (3.1) always holds for suitable constants $\eta \geq \xi > 0$.

In Section 5 we use Theorem 5 to obtain multiplicativity factors for C-numerical radii with Hermitian C.

4. Some Combinatorial Inequalities

Let α_j , γ_j , $1 \le j \le n$, be scalars and consider the set

$$\mathbf{g}_{\gamma}(\alpha) = \left\{ \sum_{j=1}^{n} \gamma_{j}^{\alpha} \sigma_{(j)} : \sigma \in \mathbf{S}_{n} \right\},$$

 $\mathbf{S}_{\mathbf{n}}$ being the symmetric group. In this section we study bounds for the radius of $\mathbf{S}_{\mathbf{v}}(\alpha)$,

$$R_{\gamma}(\alpha) = \max\{|z| : z \in g_{\gamma}(\alpha)\}.$$

A general remark is that all the involved quantities are invariant under rearrangements of the $\alpha_{\bf j}$ and the $\gamma_{\bf j}$, and under rotations of the form

$$(\alpha_1, \dots, \alpha_n) \rightarrow e^{i\phi}(\alpha_1, \dots, \alpha_n), (\gamma_1, \dots, \gamma_n) \rightarrow e^{i\psi}(\gamma_1, \dots, \gamma_n)$$

which include, of course, change of sign. This fact will be repeatedly used in the proof of the following results.

LEMMA 4. For any
$$\alpha_{\mathbf{j}}, \gamma_{\mathbf{j}} \in \mathbb{C}$$
,
$$\mathbb{R}_{\gamma}(\alpha) \geq \frac{1}{n} \begin{vmatrix} \sum_{j=1}^{n} \alpha_{j} & \sum_{j=1}^{n} \gamma_{\mathbf{i}} \\ \end{bmatrix}.$$

<u>Proof.</u> Let τ^{i} , i = 1, 2, ..., n, be the powers of a nontrivial cyclic permutation in S_{n} . Since

$$\sum_{j=1}^{n} \gamma_{j}^{\alpha} \tau^{i}(j) \in \mathfrak{A}_{\gamma}(\alpha), \quad 1 \leq i \leq n,$$

then

$$R_{\gamma}(\alpha) \geq \left| \frac{1}{n} \sum_{i=1}^{n} \left(\sum_{j=1}^{n} \gamma_{j}^{\alpha}_{\tau^{i}(j)} \right) \right|$$

$$= \left| \frac{1}{n} \sum_{j} \gamma_{j} \sum_{i} \alpha_{\tau^{i}(j)} \right| = \left| \frac{1}{n} \sum_{j} \gamma_{j} \sum_{i} \alpha_{i} \right|,$$

and the lemma holds.

LEMMA 5. If
$$\alpha_i \in \mathbb{R}$$
, $\gamma_j \in \mathbb{Q}$, $1 \le j \le n$, then
$$\mathbb{R}_{\gamma}(\alpha) \ge \frac{1}{2} \max_{i,j} |\alpha_i - \alpha_j| \max_{i,j} |\gamma_i - \gamma_j|.$$

Proof. Setting

$$\gamma_j = \lambda_j + i\mu_j, \quad \lambda_j, \mu_j \in \mathbb{R},$$

we have

$$R_{\gamma}(\alpha) = \max_{\sigma \in S_{n}} \left| \sum_{j} \lambda_{j}^{\alpha} \sigma(j) + i \sum_{j} \mu_{j}^{\alpha} \sigma(j) \right|$$

$$\geq \max_{\sigma \in S_{n}} \left| \sum_{j} \lambda_{j}^{\alpha} \sigma(j) \right| = R_{\lambda}(\alpha)$$

Now, if the γ_j are equal, then the result is trivial; so by rotating and rearanging the γ_j , we may assume that

$$\max |\gamma_i - \gamma_j| = \gamma_1 - \gamma_n > 0.$$

It follows that

$$\lambda_1 - \lambda_n = \gamma_1 - \gamma_n = \max_{i,j} |\gamma_i - \gamma_j| \ge \max_{i,j} |\lambda_i - \lambda_j|.$$

Thus

$$\lambda_1 \ge \lambda_j \ge \lambda_n$$
, $2 \le j \le n - 1$,

'so we may assume that

$$\lambda_1 \geq \lambda_2 \geq \cdots \geq \lambda_n$$

We may also assume that

$$\alpha_1 \ge \alpha_2 \ge \cdots \ge \alpha_n$$
.

Hence, observing that

$$s_1 = \sum \lambda_j \alpha_j$$
, $s_2 = \sum \lambda_j \alpha_{n-j}$

are two points in $g_{\mathbf{v}}(\alpha)$, we have

$$\begin{split} & \mathbb{R}_{\lambda}(\alpha) \geq \frac{1}{2} \Big| \mathbf{s}_{1} - \mathbf{s}_{2} \Big| = \frac{1}{2} \Big| \lambda_{1}(\alpha_{1} - \alpha_{n}) + \lambda_{2}(\alpha_{2} - \alpha_{n-1}) + \cdots + \lambda_{n}(\alpha_{n} - \alpha_{1}) \Big| \\ & = \frac{1}{2} \Big| (\lambda_{1} - \lambda_{n})(\alpha_{1} - \alpha_{n}) + (\lambda_{2} - \lambda_{n-1})(\alpha_{2} - \alpha_{n-1}) + \cdots + (\lambda_{\left[\frac{n}{2}\right]} - \lambda_{\left[\frac{n}{2}\right]} + 1)(\alpha_{\left[\frac{n}{2}\right]} - \alpha_{\left[\frac{n}{2}\right]} + 1) \Big| \\ & \geq \frac{1}{2} (\lambda_{1} - \lambda_{n})(\alpha_{1} - \alpha_{n}) = \frac{1}{2} \max |\gamma_{1} - \gamma_{1}| \max |\alpha_{1} - \alpha_{1}|, \end{split}$$

and the lemma follows.

We are interested now in obtaining constants K_{γ} , which may depend on the γ_j but not on the α_j , such that

$$(4.1) M_{\gamma}(\alpha) \geq K_{\gamma} \max |\alpha_{j}| \forall \alpha_{1}, \dots, \alpha_{n} \in \mathbb{R}.$$

THEOREM 6. For given $\gamma_j \in \mathbb{C}$, $1 \le j \le n$, there exists a constant $K_{\gamma} > 0$ which satisfies (4.1) if and only if

(4.2)
$$\gamma_j$$
 are not all equal and $\sum_j \gamma_j \neq 0$.

If (4.2) holds, then (4.1) is satisfied by the positive constant

(4.3)
$$K_{\gamma} = \frac{\begin{vmatrix} \sum \gamma_{j} \mid \cdot \max \mid \gamma_{i} - \gamma_{j} \mid}{1, j} \\ \frac{1}{2 \mid \sum \gamma_{j} \mid + \max \mid \gamma_{i} - \gamma_{j} \mid}{1, j} \end{vmatrix}$$

<u>Proof.</u> Suppose (4.2) is violated. If the γ_j are equal, we choose α_j not all equal, with $\Sigma \alpha_j = 0$; if $\Sigma \gamma_j = 0$, we take $\alpha_j = 1$, $1 \le j \le n$. In both cases $R_{\gamma}(\alpha) = 0$ but $\max |\alpha_j| > 0$; hence no positive K_{γ} satisfies (4.1).

Conversely, let (4.2) hold, and let K, be the constant specified

in (4.3). We may assume that

$$\alpha_1 \geq \cdots \geq \alpha_n$$

where in fact, by change of sign if necessary, it suffices to consider the cases

$$\alpha_1 \ge \cdots \ge \alpha_n \ge 0,$$

and

(4.4b)
$$\alpha_1 \ge \cdots \ge \alpha_k \ge 0 > \alpha_{k+1} \ge \cdots \ge \alpha_n \text{ with } \max |\alpha_j| = \alpha_1.$$

In case (4.4a) we write $\alpha_n = 6\alpha_1$, $0 \le \theta \le 1$, and use Lemmas 4 and 5 to obtain, respectively,

$$R_{\gamma}(\alpha) \geq \frac{1}{n} |\sum \alpha_{j}| |\sum \gamma_{j}| > \alpha_{n} |\sum \gamma_{j}| = \omega_{1} |\sum \alpha_{j}| = \Theta |\sum \gamma_{j}| \max |\alpha_{j}|$$

and

$$\begin{split} \mathbf{R}_{\gamma}(\alpha) &\geq \frac{1}{2} \max |\alpha_{\mathbf{i}} - \alpha_{\mathbf{j}}| \max |\gamma_{\mathbf{i}} - \gamma_{\mathbf{j}}| = \frac{1}{2} (\alpha_{\mathbf{i}} - \alpha_{\mathbf{n}}) \max |\gamma_{\mathbf{i}} - \gamma_{\mathbf{j}}| \\ &\geq \frac{1}{2} (1 - \theta) \max |\gamma_{\mathbf{i}} - \gamma_{\mathbf{j}}| \max |\alpha_{\mathbf{j}}|. \end{split}$$

We thus find that

$$R_{\gamma}(\alpha) \geq \max \left\{ \theta | \sum \gamma_{\mathbf{j}} |, \frac{1}{2} (1 - \theta) \max | \gamma_{\mathbf{j}} - \gamma_{\mathbf{j}} | \right\} \cdot \max |\alpha_{\mathbf{j}}|.$$

The expressions in the above braces are functions of θ which describe straight lines with opposite slopes and intersection value K_{γ} . Hence, for any θ

$$\max \left\{ \Theta | \sum \gamma_{\mathbf{j}} |, \frac{1}{2} (1 - \Theta) \max | \gamma_{\mathbf{i}} - \gamma_{\mathbf{j}} | \right\} \geq K_{\gamma},$$

and (4.1) follows.

In case (4.4b) we use Lemma 5 to find that

$$\mathbf{R}_{\gamma}(\alpha) \geq \frac{1}{2}(\alpha_{\mathbf{1}} - \alpha_{\mathbf{n}}) \max |\gamma_{\mathbf{1}} - \gamma_{\mathbf{n}}| > \frac{1}{2} \max |\gamma_{\mathbf{i}} - \gamma_{\mathbf{j}}| \max |\alpha_{\mathbf{j}}|.$$

Since

$$\frac{1}{2} \max |\gamma_i - \gamma_j| > K_{\gamma'}$$

then (4.1) holds again, and the theorem is proven.

The above result can be improved for certain classes of γ_{j} .

THEOREM 7. If γ_j , $1 \le j \le n$, are complex scalars of the same argument, then (4.1) holds with

(4.5)
$$K_{\gamma} = \frac{1}{2} \max_{i,j} |\gamma_i - \gamma_j|.$$

Proof. By change of argument and rearangement we may assume that

$$\gamma_1 \geq \cdots \geq \gamma_n \geq 0$$
,

and that the α_j satisfy (4.4a) or (4.4b).

For (4.4a) we have

$$R_{\gamma}(\alpha) = \sum \gamma_{j}^{\alpha} \alpha_{j} \geq \gamma_{1}^{\alpha} \alpha_{1} \geq \frac{1}{2} (\gamma_{1} - \gamma_{n}) \alpha_{1};$$

and for (4.4b), Lemma 5 yeilds

$$R_{\gamma}(\alpha) \geq \frac{1}{2}(\gamma_1 - \gamma_n)(\alpha_1 - \alpha_n) > \frac{1}{2}(\gamma_1 - \gamma_n)\alpha_1$$

Thus,

$$R_{\gamma}(\alpha) \geq \frac{1}{2} \max |\gamma_i - \gamma_j| \max |\alpha_j|,$$

and the proof is complete.

Indeed, comparing K_{γ} of (4.5) with K_{γ} of (4.3), we realize that for the relevant γ_{j} , Theorem 6 provides a tighter lower bound for $R_{\gamma}(\alpha)$ than Theorem 5.

5. Multiplicative Hermitian Radii

As indicated previously, the purpose of this section is to obtain multiplicativity factors for C-numerical radii with Hermitian C.

LEMMA 6. Let A, C be normal matrices with eigenvalues α_j and γ_j , respectively. Then

$$r_{C}(A) = R_{\gamma}(\alpha)$$
.

Proof. Obviously, it suffices to show that

conv
$$W_{C}(A) = conv S_{\gamma}(\alpha)$$
.

Since $W_C(A)$ is invariant under unitary similarities of A and C, and since A and C are normal, then by (1.1),

$$W_{C}(A) = \left\{ \sum_{j=1}^{n} \gamma_{j} x_{j}^{*} \operatorname{diag}(\alpha_{1}, \dots, \alpha_{n}) x_{j} : \{x_{j}\} \in \Lambda_{n} \right\}.$$

Thus, using the standard basis $\{e_j\}$, we find that every point in $\mathbf{g}_{\gamma}(\alpha)$ satisfies

$$\sum_{\mathbf{j}} \gamma_{\mathbf{j}}^{\alpha} \sigma(\mathbf{j}) = \sum_{\mathbf{j}} \gamma_{\mathbf{j}}^{*} e_{\sigma(\mathbf{j})}^{*} \operatorname{diag}(\alpha_{\mathbf{j}}, \dots, \alpha_{\mathbf{n}}) e_{\sigma(\mathbf{j})} \in W_{\mathbf{C}}(\mathbf{A}),$$

which gives us

$$\mathbf{g}_{\mathbf{v}}(\alpha) \subseteq \mathbf{W}_{\mathbf{C}}(\mathbf{A})$$
.

Conversely, take an arbitrary point,

$$\sum_{j} \gamma_{j} x_{j}^{*} \operatorname{diag}(\alpha_{1}, \ldots, \alpha_{n}) x_{j} \in W_{C}(A).$$

Since $x_j = (x_{j1}, \dots, x_{jn})^T$, $1 \le j \le n$, is an orthonormal basis, then $X = [|x_{jk}|^2]$ is a doubly stochastic matrix. Doubly stochastic matrices are convex combinations of permutation matrices P_{σ} . Thus writing $X = \sum_{\sigma} \lambda_{\sigma} P_{\sigma}$ and

$$a \equiv (\alpha_1, \dots, \alpha_n)^T, \quad c \equiv (\gamma_1, \dots, \gamma_n)^T,$$

we have

$$\sum_{\mathbf{j}} \gamma_{\mathbf{j}} \mathbf{x}_{\mathbf{j}}^{*} \operatorname{diag}(\alpha_{1}, \dots, \alpha_{n}) \mathbf{x}_{\mathbf{j}} = \sum_{\mathbf{j}, \mathbf{k}} \gamma_{\mathbf{j}} |\chi_{\mathbf{j} \mathbf{k}}|^{2} \alpha_{\mathbf{k}} = c^{T} \mathbf{x}_{\mathbf{a}} =$$

$$= c^{T} \left[\sum_{\sigma \in \mathbf{S}_{n}} \lambda \alpha_{\sigma} P_{\sigma} \right] \mathbf{a} = \sum_{\sigma} \lambda_{\sigma} (c^{T} P_{\sigma} \mathbf{a}) = \sum_{\sigma} \lambda_{\sigma} \left[\sum_{\mathbf{j}} \gamma_{\mathbf{j}} \alpha_{\sigma(\mathbf{j})} \right] \in \operatorname{conv} \mathbf{S}_{\gamma}(\alpha).$$

This yields

$$W_{C}(A) \subseteq \text{conv } g_{\gamma}(\alpha),$$

and the lemma follows.

LEMMA 7. Let C be normal with eigenvalues γ_j , let K_{γ} satisfy (4.1), and let

$$\|A\|_2 = \max\{(x^*A^*Ax)^{1/2} : x^*x = 1\}$$

denote the spectral norm of A. Then

$$\mathbf{r}_{\mathbf{C}}(\mathbf{A}) \geq \mathbf{K}_{\gamma} \|\mathbf{H}\|_{2} \quad \forall \underline{\mathbf{Hermitian}} \; \mathbf{H} \in \mathbf{C}_{\mathbf{n} \times \mathbf{n}}.$$

<u>Proof.</u> For Hermitian H with eigenvalues α_1 , we know that

$$\|\mathbf{H}\|_2 = \max |\alpha_j|$$
.

Since the α_{j} are real, we may use (4.1), and by Lemma 6

$$r_{C}(A) = R_{\gamma}(\alpha) \ge K_{\gamma} \max |\alpha_{j}| = K_{\gamma} ||A||_{2}.$$

LEMMA 8. If C is Hermitian, then $r_C(A) = r_C(A^*)$.

Proof.

$$r_{C}(A) = \max_{U} |tr(CU^{*}AU)| = \max_{U} |tr(CU^{*}AU)^{*}| = \max_{U} |tr(U^{*}A^{*}UC)| = r_{C}(A^{*}).$$

LEMMA 9. If C is Hermitian with eigenvalues γ_j , and if K_{γ} satisfies (4.1), then

$$r_{C}(A) \geq \frac{1}{2} K_{\gamma} ||A||_{2} \quad \forall A \in C_{n \times n}.$$

<u>Proof.</u> We write $A = \frac{1}{2}(H_1 - iH_2)$, where

$$H_1 = A + A^*, \quad H_2 = i(A - A^*),$$

are Hermitian. By Lemmas 7 and 8, and by Theorem 1,

$$\frac{1}{2} K_{\gamma} \|A\|_{2} = \frac{1}{4} K_{\gamma} \|H_{1} - iH_{2}\|_{2} \le \frac{1}{4} K_{\gamma} \|H_{1}\|_{2} + \|H_{2}\|_{2} \le \frac{1}{4} [r_{C}(H_{1}) + r_{C}(H_{2})]$$

$$= \frac{1}{4} [r_{C}(A + A^{*}) + r_{C}(A - A^{*})] \le \frac{1}{2} [r_{C}(A) + r_{C}(A^{*})] = r_{C}(A),$$

and the proof is complete.

LEMMA 10. If C is normal with eigenvalues γ_j , then

$$\mathbf{r}_{\mathbf{C}}(\mathbf{A}) \leq \sum_{\mathbf{j}} |\gamma_{\mathbf{j}}| \|\mathbf{A}\|_{2} \quad \forall \mathbf{A} \in \mathbb{Q}_{\mathbf{n} \times \mathbf{n}}.$$

Proof. By (1.1) we have

$$\mathbf{r}_{\mathbf{C}}(\mathbf{A}) = \max \left\{ \left| \sum_{\mathbf{j}} \gamma_{\mathbf{j}} \mathbf{x}_{\mathbf{j}}^* \mathbf{A} \mathbf{x}_{\mathbf{j}} \right| : \{\mathbf{x}_{\mathbf{j}}\} \in \Lambda_{\mathbf{n}} \right\};$$

and since $|x^*Ax| \le |A|_2$ for any unit vector x, the lemma follows.

THEOREM 8. Let C be Hermitian, nonscalar, with tr C \neq 0 and eigenvalues γ_j . Then, for any ν with

$$\nu \geq 4 \sum_{\mathbf{j}} |\gamma_{\mathbf{j}}| \left[\frac{2|\sum \gamma_{\mathbf{j}}| + \max_{\mathbf{i}, \mathbf{j}} |\gamma_{\mathbf{i}} - \gamma_{\mathbf{j}}|}{|\sum \gamma_{\mathbf{j}}| \cdot \max_{\mathbf{i}, \mathbf{j}} |\gamma_{\mathbf{i}} - \gamma_{\mathbf{j}}|} \right]^{2},$$

the (Hermitian) numerical radius $vr_C = r_{vC}$ is a matrix norm.

<u>Proof.</u> Since C is nonscalar, the γ_j are not all equal; and since tr C \neq 0, then $\Sigma \gamma_j \neq$ 0. Thus, by Theorem 6, the inequality in (4.1) is satisfied by the <u>positive</u> constant K_{γ} of (4.3). By Lemmas 9 and 10 we have therefore,

$$\frac{1}{2} \frac{\left| \sum \gamma_{j} \right| + \max \left| \gamma_{i} - \gamma_{j} \right|}{2\left| \sum \gamma_{j} \right| + \max \left| \gamma_{i} - \gamma_{j} \right|} \left\| A \right\|_{2} \leq r_{C}(A) \leq \sum \left| \gamma_{j} \right| \left\| A \right\|_{2} \quad \forall A \in \mathbb{Q}_{n \times n},$$

and Theorem 5 completes the proof.

For Hermitian definite C, we improve Theorem 6 as follows.

THEOREM 9. Let C be Hermitian nonnegative (nonpositive) definite.

If C is nonscalar with eigenvalues γ_j , then for every ν with

$$\nu \geq \frac{16\sum_{\mathbf{j}} |\gamma_{\mathbf{j}}|}{\max_{\mathbf{j}} |\gamma_{\mathbf{j}} - \gamma_{\mathbf{j}}|},$$

vr_C ≡ r_{vC} is a matrix norm.

<u>Proof.</u> Since C is Hermitian definite, the γ_j are of the same sign and by Theorem 7, K_{γ} of (4.5) satisfies (4.1). Lemmas 9 and 10 yield now,

$$(5.1) \quad \frac{1}{4} \max |\gamma_i - \gamma_j| \|A\|_2 \le r_C(A) \le \sum |\gamma_j| \|A\|_2 \quad \forall A \in \mathbb{Q}_{n \times n}.$$

Since C is nonscalar, the γ_j are not all equal; so $\max |\gamma_i - \gamma_j| > 0$, and Theorem 5 completes the proof.

EXAMPLE 2. We recall the definition of the k-numerical radius r_k . By Theorem 7, we find that vr_k , $1 \le k \le n - 1$, is a matrix norm if $v \ge 16k$.

Example 2 implies that $\nu \ge 16$ is a multiplicativity factor for the classical radius r. The optimal factor, ν_{r} , is given in the following result.

THEOREM 10. vr is a matrix norm if and only if $v \ge 4$; that is $v_r = 4$.

Proof. It is well known, [3, \$162], that

$$\frac{1}{2}\|\mathbf{A}\|_{2} \leq \mathbf{r}(\mathbf{A}) \leq \|\mathbf{A}\|_{2} \quad \forall \mathbf{A} \in \mathcal{C}_{n \times n}.$$

Thus, by Theorem 5, $v \ge 4$ is a multiplicativity factor for r, and by Theorem 4, $v_r \le 4$.

To show that $\nu_r \ge 4$, consider the matrices

$$A = \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} \oplus I_{n-2}, \qquad B = \begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix} \oplus I_{n-2}.$$

A simple calculation shows that

$$r(A) = r(B) = \frac{1}{2}, \quad r(AB) = 1.$$

Hence

$$r_{\nu}(AB) \leq r_{\nu}(A)r_{\nu}(B)$$

if and only if $v \ge 4$, and the theorem follows.

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Given n × n complex matrices A, B, the C-numerical radius of A is the nonnegative quantity

 $r_{C}(A) \equiv \max\{|tr(CU^*AU)| : U \text{ unitary}\}.$

For C = diag(1,0,...,0) it reduces to the classical numerical radius

(con't) 20. Abstract $r(A) = \max\{|\mathbf{x}^*A\mathbf{x}| : \mathbf{x}^*\mathbf{x} = 1\}$. We show that r_C is a generalized matrix norm if and only if C is nonscalar and tr $C \neq 0$. Next, we consider an arbitrary generalized matrix norm and characterize all constants v > 0 for which v is multiplicative. A technique to obtain such v is then applied to C-numerical radii with Hermitian C. In particular we find that vr is a matrix norm if and only if $v \geq 4$.